

# Stochastic differential equations

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# Ordinary Differential Equations (ODE's)

- An ODE consists in a functional relation between a function and its derivatives.

Example: A  $k$ th order ODE in implicit form

$$a\left(x(t), x'(t), x''(t), \dots, x^{(k)}(t)\right) = 0$$

- A first order ODE in explicit form is

$$\frac{dx(t)}{dt} = a(t, x(t))$$

# Formal solution of an ODE

- Consider the first order ODE in explicit form

$$\frac{dx(t)}{dt} = a(t, x(t))$$

- The solution of this ODE for an initial condition  $x(t_0) = x_0$  is

$$x(t) = x(t_0) + \int_{t_0}^t a(s, x(s)) ds$$

# Closed-form solution of a simple ODE

- Consider the first order ODE

$$\frac{dx(t)}{dt} = r x(t)$$

- The solution of this ODE for an initial condition  $x(t_0) = x_0$  is

$$x(t) = x_0 \exp\{r(t - t_0)\}$$

# Numerical solution of an ODE

- Consider the alternative form of a first order ODE

$$dx(t) = a(t, x(t))dt$$

- Using the definition of the differential:

$$dx(t) = x(t + dt) - x(t)$$

- It is possible to write

$$x(t + dt) = x(t) + a(t, x(t))dt$$

# Numerical solution of an ODE: Euler scheme

- Initial condition:  $x(t_0) = x_0$
- Integration interval  $[t_0, t_0 + T]$
- Grid of points for integration:  $t_0 < t_1 < \dots < t_N = t_0 + T$ ;
- Solution trajectory:

$$x_{n+1} = x_n + a(t_n, x_n)\Delta t_n; \quad n = 0, 1, \dots, (N - 1)$$

$$x_n \stackrel{\text{def}}{=} x(t_n); \quad \Delta t_n \stackrel{\text{def}}{=} t_{n+1} - t_n \longrightarrow 0^+$$

# Euler scheme for a simple ODE

- Consider the first order ODE

$$\frac{dx(t)}{dt} = rx(t);$$

- The numerical solution of this ODE starting from the i. c.  $x(t_0) = x_0$

$$x_{n+1} = x_n(1 + r\Delta t_n); \quad n = 0, 1, \dots, (N - 1)$$

- Solution at  $t_N = t_0 + T$

$$x_N = x_0 \prod_{n=0}^{N-1} (1 + r\Delta t_n) \xrightarrow{\Delta t_n \rightarrow 0^+} x_0 \exp\left(r \sum_{n=0}^{N-1} \Delta t_n\right) = x_0 \exp(rT)$$

# Euler scheme: Regular grid

- Initial condition:  $x(t_0) = x_0$
- Integration interval  $[t_0, t_0 + T]$
- Regular grid of integration points :  $t_n = t_0 + n\Delta T; \quad n = 0, 1, \dots, N$
- Solution trajectory:

$$x_{n+1} = x_n + a(t_n, x_n)\Delta T; \quad n = 0, 1, \dots, (N - 1)$$

$$x_n \stackrel{\text{def}}{=} x(t_n); \quad \Delta T \stackrel{\text{def}}{=} \frac{T}{N} \longrightarrow 0^+$$



# Stochastic Differential Equation (SDE)

- Consider the first order ODE in explicit form

$$dX(t) = a(t, X(t))dt + b(t, X(t))dW(t)$$

where  $W(t)$  is the Wiener process (standard Brownian motion).

- The formal solution of this equation, assuming that the process starts at the initial condition  $X(t_0) = x_0$  with probability one, is

$$X(t) = x_0 + \int_{t_0}^t a(s, X(s)) ds + \int_{t_0}^t b(s, X(s)) dW(s)$$

# Non-anticipating functions

Related to causality:  
The present ( $t$ )  
cannot be affected by  
the future ( $t + \tau$ )

- The function  $g(t, \omega)$  is non-anticipating if

$$\forall t, \tau > 0 : \quad g(t, \omega) \perp (W(t + \tau, \omega) - W(t, \omega))$$

- Examples of non-anticipating functions

- $W(t, \omega), \quad \int_{t_0}^t f(W(s, \omega)) ds, \quad \int_{t_0}^t f(W(s, \omega)) dW(s, \omega)$

- Assuming that  $g(t, \omega)$  is non-anticipating,

the functions  $\int_{t_0}^t g(s, \omega) ds, \int_{t_0}^t g(s, \omega) dW(s, \omega)$  are also non-anticipating.

# Stochastic Integral (Itô)

- Consider the stochastic integral

$$I(\omega) = \int_{t_0}^t g(s, \omega) dW(s, \omega)$$

Non-anticipating  
function

$\omega \in \Omega$  indicates that it is a particular realization (trajectory) of the Wiener process. We will omit it when there is no ambiguity as to the source of randomness.

- Consider the Riemann sum in the grid  $t_0 < t_1 < \dots < t_N = t$

$$I_N(\omega) = \sum_{n=0}^{N-1} g(t_n, \omega) (W(t_{n+1}, \omega) - W(t_n, \omega))$$

The time at which  $g$  is evaluated matters!

- The stochastic integral is the limit  $N \rightarrow \infty$  of this Riemann sum

Convergence in the mean square sense:

$$\lim_{N \rightarrow \infty} \mathbb{E}[(I_N - I)^2] = 0$$

$$I(\omega) = \lim_{N \rightarrow \infty} I_N(\omega)$$

It is a random variable! Its value depends on the particular realization of the random process

# A simple stochastic integral

- Consider the stochastic (Itô) integral

$$I = \int_{t_0}^t c \, dW(s)$$

- Applying the definition of the stochastic integral

$$\begin{aligned} I &= \lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} c \left( W(t_{n+1}) - W(t_n) \right) = c \lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} \left( W(t_{n+1}) - W(t_n) \right) \\ &= c(W(t) - W(t_0)) \end{aligned}$$

# A more complex stochastic integral

- Consider the stochastic (Itô) integral

$$S = \int_{t_0}^t W(s) dW(s)$$

- Applying the definition of the stochastic integral

$$\begin{aligned} S &= \lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} W(t_n) (W(t_{n+1}) - W(t_n)) \\ &= \frac{1}{2} \lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} \left[ (W(t_{n+1})^2 - W(t_n)^2) - (W(t_{n+1}) - W(t_n))^2 \right] \\ &= \frac{1}{2} (W(t)^2 - W(t_0)^2) - \frac{1}{2} \lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} [(t_{n+1} - t_n) Z_n^2] \\ &= \frac{1}{2} (W(t)^2 - W(t_0)^2) - \frac{1}{2} (t - t_0) \end{aligned}$$

$$\begin{aligned} &\mathbb{E} \left[ \lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} [(t_{n+1} - t_n) Z_n^2] \right] \\ &= \lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} [(t_{n+1} - t_n) \mathbb{E}[Z_n^2]] \\ &= (t - t_0) \\ \text{Var} \left[ \lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} [(t_{n+1} - t_n) Z_n^2] \right] &= \lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} [(t_{n+1} - t_n)^2 \text{Var}[Z_n^2]] \\ &\leq 2 \max_n (t_{n+1} - t_n) \lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} (t_{n+1} - t_n) \\ &= 2 \max_n (t_{n+1} - t_n) (t - t_0) \end{aligned}$$

# Properties of Itô integrals

- Consider the non-anticipating functions  $g(s, \omega), g_1(s, \omega), g_2(s, \omega)$ 
  - $\mathbb{E} \left[ \int_{t_0}^t g(s, \omega) dW(s, \omega) \right] = 0$
  - $\text{Var} \left[ \int_{t_0}^t g(s, \omega) dW(s, \omega) \right] = \int_{t_0}^t \mathbb{E}[g^2(s, \omega)] ds < \infty$   
(condition for the Itô stochastic integral to exist)
  - $\mathbb{E} \left[ \int_{t_0}^t g_1(s, \omega) dW(s, \omega) \int_{t_0}^t g_2(s, \omega) dW(s, \omega) \right] = \int_{t_0}^t \mathbb{E}[g_1(s, \omega)g_2(s, \omega)] ds$
  - Linearity:
$$\int_{t_0}^t [a_1 g_1(s, \omega) + a_2 g_2(s, \omega)] dW(s, \omega)$$
$$= a_1 \int_{t_0}^t g_1(s, \omega) dW(s, \omega) + a_2 \int_{t_0}^t g_2(s, \omega) dW(s, \omega)$$

$$dW^2(t) = dt$$

- Consider the non-anticipating, bounded function  $g(t, \omega)$ , then

$$\int_{t_0}^t g(s, \omega) [dW(s, \omega)]^2 = \int_{t_0}^t g(s, \omega) ds$$

$$dW^{2+n}(t) = 0; \quad n > 0$$

- Consider the non-anticipating, bounded function  $g(t, \omega)$ , then

$$\int_{t_0}^t g(s, \omega) [dW(s, \omega)]^{2+n} = 0$$



# General differentiation rules

- $[dW(t)]^2 \rightarrow dt$
- $[dW(t)]^{2+n} \rightarrow 0 \quad (n > 0)$
- $dt \, dW(t) \rightarrow 0$
- $(dt)^{1+n} \rightarrow 0 \quad (n > 0)$
- All higher powers vanish
- $\mathbb{E} [dW(t, \omega) \, dW(t', \omega)] \rightarrow \delta(t - t') \, dt \, dt'$

$$d\varphi(t, W(t)) = \left( \varphi_t(t, W(t)) + \frac{1}{2} \varphi_{WW}(t, W(t)) \right) dt + \varphi_W(t, W(t)) dW(t)$$

# Itô's lemma

- Consider the SDE

$$dX(t) = a(t, X(t))dt + b(t, X(t))dW(t)$$

- The SDE for  $Y(t) = \varphi(t, X(t))$  is

$$dY(t) = \tilde{a}(t, X(t))dt + \tilde{b}(t, X(t))dW(t)$$

$$\tilde{a}(t, X(t)) = \varphi_t(t, X(t)) + \varphi_x(t, X(t))a(t, X(t)) + \frac{1}{2} \varphi_{xx}(t, X(t)) b(t, X(t))^2$$

$$\tilde{b}(t, X(t)) = \varphi_x(t, X(t))b(t, X(t))$$

# General differentiation rules: Several variables

- $dW_i(t)$  behaves as a differential  $(dt)^{1/2}$ .
- $dW_i(t)dW_j(t) \rightarrow \delta_{ij}dt; \quad i, j = 1, 2, \dots, D$
- $[dW_i(t)]^{2+n} \rightarrow 0 \quad (n > 0)$
- $dt dW_i(t) \rightarrow 0$
- $(dt)^{1+n} \rightarrow 0 \quad (n > 0)$
- All higher powers vanish.

# Itô's lemma: Geometric Brownian Motion

- Consider the SDE

$$dS(t) = \mu S(t)dt + \sigma S(t) dW(t)$$

- Change of variable:  $Y(t) = \log S(t)$

- $\varphi(t, S(t)) = \log S(t)$

- $\varphi_t(t, S(t)) = 0; \quad \varphi_S(t, S(t)) = \frac{1}{S(t)}; \quad \varphi_{SS}(t, S(t)) = -\frac{1}{S(t)^2}$

$$dY(t) = \left( \mu - \frac{1}{2} \sigma^2 \right) dt + \sigma dW(t)$$

# Geometric Brownian Motion SDE: Solution

- Consider the SDE:  $dY(t) = \left(\mu - \frac{1}{2}\sigma^2\right) dt + \sigma dW(t)$

$$Y(t) = Y(t_0) + \left(\mu - \frac{1}{2}\sigma^2\right)(t - t_0) + \sigma(W(t) - W(t_0))$$

- Undoing the change of variable:  $Y(t) = \log S(t)$

$$S(t) = S(t_0) \exp \left\{ \left(\mu - \frac{1}{2}\sigma^2\right)(t - t_0) + \sigma(W(t) - W(t_0)) \right\}$$

# Numerical solution of an SDE: Stochastic Euler scheme (order 1/2)

- Initial condition:  $x(t_0) = x_0$
- Integration interval  $[t_0, t_0 + T]$
- Grid of points for integration:  $t_0 < t_1 < \dots < t_N = t_0 + T$ ;
- Solution:  $M \rightarrow \infty$  trajectories

$$x_0^{(m)} = x_0; \quad m = 1, \dots, M.$$

$$x_{n+1}^{(m)} = x_n^{(m)} + a(t_n, x_n^{(m)}) \Delta t_n + b(t_n, x_n^{(m)}) \sqrt{\Delta t_n} Z_n^{(m)} \\ n = 0, 1, \dots, (N-1); \quad m = 1, \dots, M.$$

$$x_n^{(m)} \stackrel{\text{def}}{=} x^{(m)}(t_n); \quad \Delta t_n \stackrel{\text{def}}{=} t_{n+1} - t_n \rightarrow 0^+; \quad \{Z_n^{(m)} \sim N(0,1)\} \text{ iidrv's}$$

Value of the  $m$ th  
trajectory at instant  $t_n$

# Stochastic Euler scheme: Regular grid

- Initial condition:  $x(t_0) = x_0$
- Integration interval  $[t_0, t_0 + T]$
- Regular grid of integration points :  $t_n = t_0 + n\Delta T; \quad n = 0, 1, \dots, N$
- Solution:  $M \rightarrow \infty$  trajectories

$$x_0^{(m)} = x_0; \quad m = 1, \dots, M.$$

$$x_{n+1}^{(m)} = x_n^{(m)} + a(t_n, x_n^{(m)})\Delta T + b(t_n, x_n^{(m)})\sqrt{\Delta T}Z_n^{(m)} \\ n = 0, 1, \dots, (N-1); \quad m = 1, \dots, M.$$

$$x_n^{(m)} \stackrel{\text{def}}{=} x^{(m)}(t_n); \quad \Delta T \stackrel{\text{def}}{=} \frac{T}{N} \rightarrow 0^+; \quad \{Z_n^{(m)} \sim N(0,1)\} \text{ iidrv's}$$

Value of the  $m$ th  
trajectory at instant  $t_n$

# Numerical solution of an SDE: Milstein scheme (order 1)

- Solution:  $M \rightarrow \infty$  trajectories

Timothy Sauer. 2013. Computational solution of stochastic differential equations. WIREs Comput. Stat. 5, 5 (September 2013), 362–371.

<http://math.gmu.edu/~tsauer/pre/wires.pdf>

$$x_0^{(m)} = x_0; \quad m = 1, \dots, M.$$

$$\begin{aligned} x_{n+1}^{(m)} = & x_n^{(m)} + a(t_n, x_n^{(m)}) \Delta t_n + b(t_n, x_n^{(m)}) \sqrt{\Delta t_n} Z_n^{(m)} \\ & + \frac{1}{2} b(t_n, x_n^{(m)}) b_x(t_n, x_n^{(m)}) \left( (Z_n^{(m)})^2 - 1 \right) \Delta t_n \\ & n = 0, 1, \dots, (N-1); \quad m = 1, \dots, M. \end{aligned}$$

$$x_n^{(m)} \stackrel{\text{def}}{=} x^{(m)}(t_n); \quad \Delta t_n \stackrel{\text{def}}{=} t_{n+1} - t_n \rightarrow 0^+; \quad \{Z_n^{(m)} \sim N(0,1)\} \text{ iidrv's}$$

Value of the  $m$ th trajectory  
at instant  $t_n$



# Fokker-Planck equation (conditional probability density)

- Consider the SDE with the initial condition  $X(t_0) = x_0$

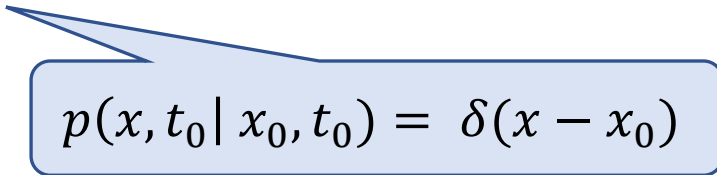
$$dX(t) = A(t, X(t))dt + \sqrt{B(t, X(t))}dW(t)$$

- The corresponding Fokker-Planck equation is

$$\partial_t p(x, t | x_0, t_0) = -\partial_x [A(t, x)p(x, t | x_0, t_0)] + \frac{1}{2} \partial_x^2 [B(t, X(t))p(x, t | x_0, t_0)]$$

Proof: Apply Itô's lemma to compute  $\frac{d}{dt} \mathbb{E}[f(X(t))]$

$$\mathbb{E}[f(X(t))] = \int f(x)p(x, t | x_0, t_0) dx$$


$$p(x, t_0 | x_0, t_0) = \delta(x - x_0)$$

# Fokker-Planck equation (general probability density)

- Assuming that the distribution at  $t_0$  is  $p_0(x)$

$$p(x, t) = \int p(x, t; x_0, t_0) dx_0 = \int p(x, t | x_0, t_0) p_0(x_0) dx_0$$

- The corresponding Fokker-Planck equation is

$$\partial_t p(x, t) = -\partial_x [A(t, x)p(x, t)] + \frac{1}{2} \partial_x^2 [B(t, x)p(x, t)]$$

$$p(x, t)|_{t=t_0} = p_0(x)$$

# Ornstein-Uhlenbeck Process (Fokker-Planck)

$k$ : rate,  
inverse correlation time

$D$ : diffusion constant

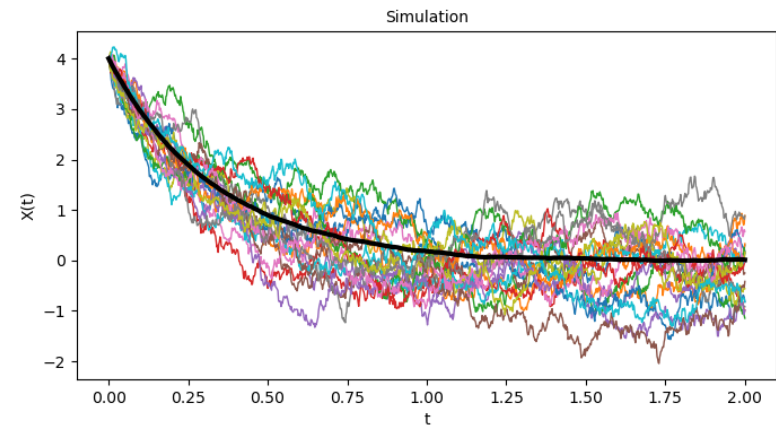
$$\partial_t p(x, t | x_0, 0) = \partial_x [k x p(x, t | x_0, 0)] + \frac{1}{2} D \partial_x^2 p(x, t | x_0, 0)$$

- Consider the equation for the characteristic function  $\phi(s, t) = \int_{-\infty}^{\infty} e^{isx} p(x, t | x_0, 0) dx$

$$\partial_t \phi(s, t) = -ks \phi(s, t) - \frac{1}{2} D s^2 \phi(s, t)$$

- The solution is

$$\phi(s, t) = \exp \left[ -\frac{Ds^2}{4k} (1 - e^{-2kt}) + isx_0 e^{-kt} \right]$$



# Stationary solution for the OU Process

- The stationary solution for the characteristic function of the OU process is

$$\phi_s(s) = \lim_{t \rightarrow \infty} \phi(s, t) = \lim_{t \rightarrow \infty} \exp \left[ -\frac{Ds^2}{4k} (1 - e^{-2kt}) + isx_0 e^{-kt} \right] = \exp \left[ -\frac{Ds^2}{4k} \right]$$

- Performing the inverse Fourier transform, we obtain the stationary solution of the OU process

$$p_s(x) = \left( \frac{k}{\pi D} \right)^{1/2} \exp \left[ -k \frac{x^2}{D} \right]$$

# Ornstein-Uhlenbeck Process (SDE)

$D$ : diffusion constant

$$dX(t) = -kX(t)dt + \sqrt{D}dW(t)$$

- Solution:  $X(t) = x_0 e^{-kt} + \sqrt{D} \int_0^t e^{-k(t-s)} dW(s)$

$k$ : rate,  
inverse correlation time

$$\mathbb{E}[X(t)] = x_0 e^{-kt}$$

$$\text{Var}[X(t)] = \frac{D}{2k} (1 - e^{-2kt})$$

$$X(t) = x_0 e^{-kt} + \sqrt{\frac{D}{2k} (1 - e^{-2kt})} Z; \quad Z \sim N(0,1)$$

$$\begin{aligned} \text{Cov}[X(t), X(t')] &= D \int_0^t \int_0^{t'} e^{-k(t+t'-s-s')} \mathbb{E}[dW(s)dW(s')] = D \int_0^{\min(t,t')} e^{-k(t+t'-2s)} ds \\ &= -\frac{D}{2k} e^{-k(t+t')} + \frac{D}{2k} e^{-k|t-t'|} \end{aligned}$$

# Simulation scheme for Ornstein Uhlenbeck

- Initial condition:  $x(t_0) = x_0$
- Integration interval  $[t_0, t_0 + T]$
- Regular grid of integration points :  $t_n = t_0 + n\Delta T; \quad n = 0, 1, \dots, N$
- Solution:  $M \rightarrow \infty$  trajectories

$$x_0^{(m)} = x_0; \quad m = 1, \dots, M.$$

$$x_{n+1}^{(m)} = x_n^{(m)} e^{-k\Delta T} + \sqrt{\frac{D}{2k} (1 - e^{-2k\Delta T})} Z_n^{(m)}$$
$$n = 0, 1, \dots, (N - 1); \quad m = 1, \dots, M.$$

$$x_n^{(m)} \stackrel{\text{def}}{=} x^{(m)}(t_n); \quad \Delta T \stackrel{\text{def}}{=} \frac{T}{N}; \quad \{Z_n^{(m)} \sim N(0,1)\} \text{ iidrv's}$$

Value of the  $m$ th  
trajectory at instant  $t_n$

# Jump-diffusion SDE

$$X(t^-) = \lim_{\substack{s \rightarrow t \\ s < t}} X(s)$$

$J(t)$  is a jump process

$$dX(t) = a(t, X(t^-))dt + b(t, X(t^-))dW(t) + c(t, X(t^-))dJ(t)$$

First jump at  $\tau_1$

Second jump at  $\tau_2$

- Jump process:

- The jumps occur at random arrival times  $t_0 < \tau_1 < \tau_2 < \dots$
- The  $j$ th jump occurs at time  $\tau_j$  and has a magnitude  $Y_j \sim f_Y(y)$ .
- Consider the counting process  $N(t) = \sum_{j=1}^{\infty} \mathbb{I}[t > \tau_j]$

$$J(t) = \sum_{j=1}^{N(t)} Y_j$$

# Jump-diffusion SDE: numerical integration

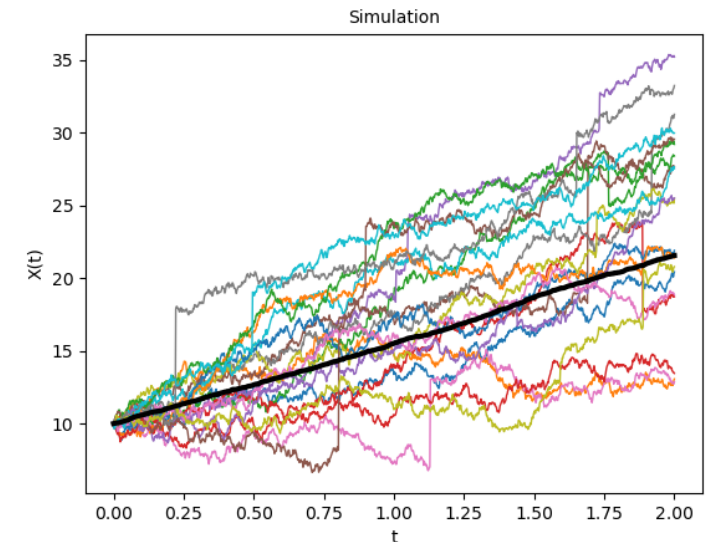
$$dX(t) = a(t, X(t^-))dt + b(t, X(t^-))dW(t) + c(t, X(t^-))dJ(t)$$

- Integration in an interval without any jump  $[t, t + \Delta T]$ ,  $\Delta T \rightarrow 0^+$

$$X(t + \Delta T) = a(t, X(t))\Delta T + b(t, X(t))\sqrt{\Delta T} Z; \quad Z \sim N(0,1)$$

- Integration in interval with a jump  $[\tau_j^-, \tau_j]$ ,

$$X(\tau_j) = X(\tau_j^-) + c(\tau_j, X(\tau_j^-)) Y_j; \quad Y_j \sim f_Y(y)$$





# Merton's jump-diffusion SDE

$$dS(t) = \mu S(t^-)dt + \sigma S(t^-)dW(t) + S(t)dJ(t)$$

iidrv's

$$S(t) = S(0)\exp\left\{\left(\mu - \frac{1}{2}\sigma^2\right)t + \sigma\sqrt{t}Z\right\}\prod_{j=1}^{N(t)}Y_j \quad ; \quad Z \sim N(0,1); Y_j \sim LN(\gamma, \delta)$$

- Integration in an interval without any jump  $[t_n, t_{n+1}]$ ,  $\Delta T_n = t_{n+1} - t_n$

$$S(t_{n+1}) = S(t_n)\exp\left\{\left(\mu - \frac{1}{2}\sigma^2\right)\Delta T_n + \sigma\sqrt{\Delta T_n}Z_n\right\}; Z_n \sim N(0,1)$$

- Integration in interval with a jump  $[\tau_j^-, \tau_j]$ ,

$$S(\tau_j) = S(\tau_j^-)Y_j; \quad Y_j \sim LN(\gamma, \delta)$$